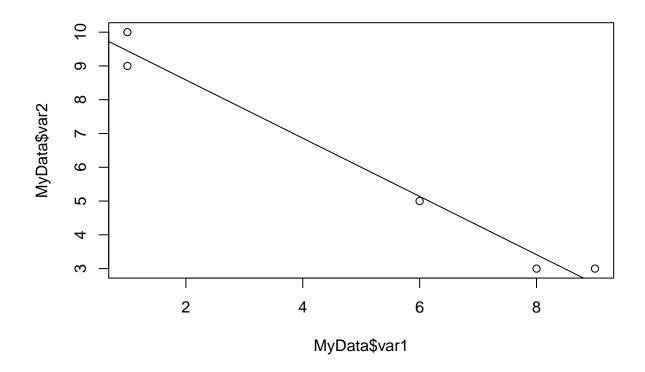
Regression Simulator. R

Administrator

Tue Apr 05 15:01:29 2016

```
# Analyzing regression data
# 5 April 2016
# NJG
# eyeball data directly from a published scatter plot.
# each observation is a paired x,y data point
obs1 <- c(1,10)
obs2 <- c(9,3)
obs3 <- c(6,5)
obs4 <- c(1,9)
obs5 <- c(8,3)
MyData <- data.frame(rbind(obs1,obs2,obs3,obs4,obs5))</pre>
colnames(MyData) <- c("var1","var2")</pre>
plot(x=MyData$var1, y=MyData$var2)
MyModel <- lm(MyData$var2~MyData$var1)</pre>
z <- summary(MyModel)</pre>
abline(summary(MyModel))
## Warning in abline(summary(MyModel)): only using the first two of 8
```

regression coefficients



extract parameters from model object print(z)

[1] "call"

```
##
## Call:
## lm(formula = MyData$var2 ~ MyData$var1)
##
## Residuals:
                                      5
##
                2
                       3
   ##
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 10.31034
                         0.43426
                                  23.74 0.000164 ***
                         0.07178 -12.01 0.001242 **
## MyData$var1 -0.86207
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
\mbox{\tt\#\#} Residual standard error: 0.5467 on 3 degrees of freedom
## Multiple R-squared: 0.9796, Adjusted R-squared: 0.9728
## F-statistic: 144.2 on 1 and 3 DF, p-value: 0.001242
names(z)
```

"residuals"

"terms"

"coefficients"

```
## [5] "aliased"
                                         "df"
                         "sigma"
                                                          "r.squared"
## [9] "adj.r.squared" "fstatistic"
                                         "cov.unscaled"
z$coefficients[c(1,2)]
## [1] 10.310345 -0.862069
# set up model parameters
intercept <- 10.31 # fitted regression intercept</pre>
slope <- -0.86 # fitted regression slope</pre>
sampleSize <- 20
                            # simulated sample size
xRange <- c(0,10) # permissable range of x-values
                 # standard deviation of residuals
residSD <- 1
RegSim <- function(a=intercept,b=slope,n=sampleSize,xl=xRange,sd=residSD,scatterplot=FALSE){</pre>
  # simulate x values from random uniform over range
 xSim <- runif(n=sampleSize,min=xRange[1],max=xRange[2])</pre>
  # simulate y values from regression plus sd for normal noise
 ySim <- a + b*xSim + rnorm(n=n,mean=0,sd=sd)
# create data frame of simulated data
SimData <- data.frame(cbind(xSim,ySim))</pre>
colnames(SimData) <- c("x","y")</pre>
# create and save summary of linear regression model
SimModel <- summary(lm(SimData$y~SimData$x))</pre>
# optionally, create regression plot
if(scatterplot==TRUE){
 plot(x=SimData$x,y=SimData$y,xlab= "Simulated X", ylab= "Simulated Y",type="n")
  grid()
  abline(SimModel,lwd=2)
  points(x=SimData$x,y=SimData$y,cex=2,pch=21,bg="wheat")
# create output as a list with data and model
output <- list(SimModel,SimData)</pre>
return(output)
}
RegSim(sd=1,scatterplot=TRUE)
## Warning in abline(SimModel, lwd = 2): only using the first two of 8
```

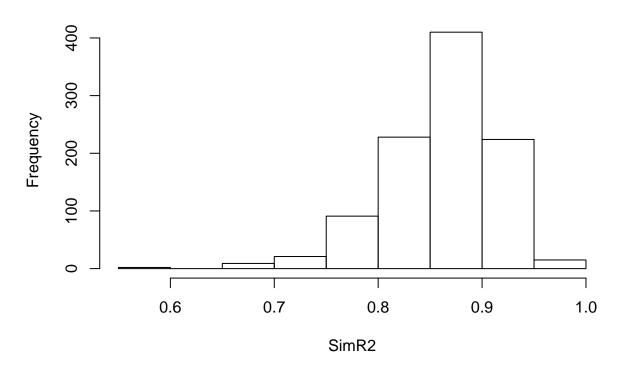
regression coefficients

```
0 2 4 6 8 10
Simulated X
```

```
## [[1]]
##
## Call:
## lm(formula = SimData$y ~ SimData$x)
##
## Residuals:
##
       Min
                  1Q
                      Median
                                   ЗQ
                                           Max
## -1.43339 -0.39893 -0.02275 0.58481 1.72382
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 10.19311
                          0.45204
                                    22.55 1.20e-14 ***
                           0.07501 -10.99 2.04e-09 ***
## SimData$x
              -0.82456
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.8549 on 18 degrees of freedom
## Multiple R-squared: 0.8703, Adjusted R-squared: 0.8631
## F-statistic: 120.8 on 1 and 18 DF, p-value: 2.044e-09
##
##
## [[2]]
##
             x
## 1 1.708133 9.938476
## 2 6.164366 5.686583
## 3 0.237257 9.224656
```

```
## 4 8.298412 3.453781
## 5 7.484622 3.831316
## 6 4.760070 5.558157
## 7 5.460008 6.569607
## 8 5.880177 7.068359
## 9 7.367821 3.842186
## 10 6.701637 5.445836
## 11 9.074515 2.636171
## 12 3.355260 8.036592
## 13 2.868279 7.856975
## 14 9.859949 2.458924
## 15 4.728406 5.998999
## 16 3.430857 5.937131
## 17 6.631865 4.466936
## 18 1.536420 9.155293
## 19 7.629394 2.468811
## 20 6.034341 4.175454
# tweak the noise in the model to get r2 = 0.44
replicates <- 1000
SimR2 <- rep(0,replicates)</pre>
for (i in 1:replicates) {
 SimR2[i] <- RegSim(sd=1,scatterplot=FALSE)[[1]]$r.squared</pre>
}
hist(SimR2)
```

Histogram of SimR2



mean(SimR2)

[1] 0.8615872

quantile(SimR2,probs=c(0.025,0.975))

2.5% 97.5% ## 0.7424873 0.9448751